	Laboratory work 1  Scebec Mihai, IS-211M  Additional sorry for making this so lame-like. It is first time I see jupyter and it is 5 minutes since I saw markdowns. So far I don't even see
In [51]:	what parts of the code should be commented more, so I am hoping I will get better at it.  # 1. imports import pandas as pd import numpy as np
In [52]:	<pre>import matplotlib.pyplot as plt  # 2. upload the dataset dataset = pd.read_csv('diabetes.csv')</pre>
In [53]:	# 3. explore dataset dataset.describe()
Out[53]:	Pregnancies         Glucose         BloodPressure         SkinThickness         Insulin         BMI         DiabetesPedigreeFunction         Age         Outcome           count         768.000000
<pre>In [54]: Out[54]:</pre>	<pre># 4. list variable names dataset.columns.tolist()  ['Pregnancies',     'Glucose',     'BloodPressure',     'SkinThickness',     'Insulin',     'BMI',     'DiabetesPedigreeFunction',     'Age',     'Outcome']</pre>
In [55]:	<pre># 5. count how many have diabetes and how many don't value_counts = dataset.Outcome.value_counts() print('Diabetes counts: ') print(str(value_counts))</pre> Diabetes counts: 0 500
In [56]:	Name: Outcome, dtype: int64  # 6. Create a bar chart to represent the frequencies of those with diabetes versus those without diabetes, # label the axes and name the chart. not_diabetes = value_counts[0] diabetes = value_counts[1] diabetes_havers_labels = ['Diabetes', 'W/ diabetes'] plt.bar(diabetes_havers_labels, [diabetes, not_diabetes]) plt.title('Frequency of having diabetes versus not having diabetes') plt.xlabel('Have a disease vs not have one') plt.ylabel('Amount of people') plt.show()  Frequency of having diabetes versus not having diabetes  500
	400 - 100 - 100 - Diabetes W/ diabetes  Have a disease vs not have one
In [57]:	# 7. Create a pie plot, change the colors, add the labels, add shade, and 'explode' the pie slice for those a "Create the legend and put it into the right corner so that it does cover the pie plot.  pie_colors = ['white', 'orange'] plt.pie(value_counts, labels=diabetes_havers_labels, colors=pie_colors, explode=[0, 0.1], shadow=True, autopo # matplotlib.rcParams['text.color'] = 'white' # for dark theme  plt.legend() plt.legend(loc="upper right") plt.subplots_adjust(left=0.01, bottom=0.001, right=0.75) plt.show()  Diabetes  W/ diabetes
	65.1% 34.9% W/ diabetes
In [58]:	<pre># 8. Calculate and report the mean the standard deviation of the BMI variable. bmi_mean = dataset.BMI.mean() bmi_std = dataset.BMI.std() print('BMI mean: ' + str(bmi_mean)) print('BMI standard deviation: ' + str(bmi_std))  BMI mean: 31.992578124999998 BMI standard deviation: 7.884160320375446  # 9. By using the Chebysheffs Theorem infer what is the range for the BMI variable for 75% of the people in y What is the range for 89% of the people in your dataset. # 1) mean - (lower boundary) = within number</pre>
	<pre># 2) mean - (upper boundary) = within number # k - number of standard deviations # k = the within number / std # wn = sqrt(1/(1-x))* std; x = percentage wn75 = np.sqrt(1/(1 - 75/100)) * bmi_std wn89 = np.sqrt(1/(1 - 89/100)) * bmi_std print(wn75) print('wn75) print('range for 75% is between '+ str(bmi_mean - wn75) + ' and ' + str(bmi_mean + wn75)) print(wn89) print('range for 89% is between '+ str(bmi_mean - wn89) + ' and ' + str(bmi_mean + wn89))  15.768320640750892 range for 75% is between 16.224257484249108 and 47.76089876575089 23.771637790630525 range for 89% is between 8.220940334369473 and 55.76421591563052</pre>
In [60]:	# 10. Create a histogram for the BMI variable, plot the mean and meadian on the histogram to answer the following the series of the BMI distribution skewed based on the position of the mean versus the median?  # b) Calculate the skewness of BMI  # c) Based on the value of BMI median and on the chart below infer what percentage of people in your dataset. BMI plt.hist(dataset.BMI)  plt.title('BMI histogram')  plt.xlabel('BMI')  plt.ylabel('Amount of people')  bmi_median = np.median(dataset.BMI)  print('BMI median = '+str(bmi_median))  plt.axvline(bmi_mean, color='yellow', linestyle='-')  plt.axvline(bmi_median, color='red', linestyle='')  plt.xshow()  # a: it doesn't seem that much skewed based on any of these variables;  # b: skewness is negative, therefore there are tails on the left side  bmi_skewness = dataset.BMI.skew()  print('BMI skewness = '+str(bmi_skewness))  # c:  # obese people are with bmi = 30 or higher  print('there is this exact percantage of people with obesity: '+ str(len(dataset.BMI[dataset.BMI] >= 30])/(length)
	BMI median = 32.0  BMI histogram  250 - 200 - 150 - 50 - 60 70
In [61]:	there is this exact percantage of people with obesity: 61.458333333333336  # 11. doing step 10 for blood pressure plt.hist(dataset.BloodPressure) plt.title('Blood pressure histogram') plt.xlabel('Blood pressure') plt.ylabel('Amount of people') blood_pressure_median = np.median(dataset.BloodPressure) blood_pressure_mean = np.mean(dataset.BloodPressure) blood_pressure_mean = np.mean(dataset.BloodPressure) print('Blood pressure median = '+str(blood pressure median)) plt.axvline(blood_pressure_mean, color='yellow', linestyle='-') plt.axvline(blood_pressure_median, color='red', linestyle='') plt.rcParams['figure.figsize'] = [10, 5] plt.show() # a: distribution is skewed on the mean more it seems # b: calculate skewness, it is negative, therefore tails are on the left, but it is also VERY big, so we show blood pressure_skewn() print('Blood pressure skewness = '+str(blood_pressure_skew()) print('Blood pressure skewness = '+str(blood_pressure_skewness)) # c: # obesity is not the thing here # I found out that with more outliers on the left the mean value became less descriptive and according to the stick to the median instead  Blood pressure median = 72.0  Blood pressure histogram  250
	250 - 200 - 20 40 60 80 100 120 Blood pressure
<pre>In [62]: Out[62]:</pre>	# 12. Remove all the observations where the BloodPressure' and the BMI variables take the value of zero dataset = dataset.drop(dataset[dataset.BloodPressure == 0].index) # for some reason it stopped working in one dataset = dataset.drop(dataset[dataset.BMI == 0].index) dataset.describe()    Pregnancies   Glucose   BloodPressure   SkinThickness   Insulin   BMI   DiabetesPedigreeFunction   Age   Outcome
In [63]:	min         0.000000         0.000000         24.000000         0.000000         0.000000         18.200000         0.078000         21.000000         0.000000           25%         1.000000         99.000000         64.000000         0.000000         0.000000         27.500000         0.245000         24.000000         0.000000           50%         3.000000         117.000000         72.000000         24.000000         46.000000         32.400000         0.378000         29.000000         0.000000           75%         6.000000         141.000000         80.000000         33.000000         130.000000         36.600000         0.627000         41.000000         1.000000           max         17.000000         199.000000         122.000000         99.000000         846.000000         67.100000         2.420000         81.000000         1.000000           #         13.8         Now perform once again all the steps in 8, 9, 10 and 11         #         8. Calculate and report the mean the standard deviation of the BMI variable.           bmi_mean = dataset.BMI.std()         print('BMI mean: ' + str(bmi mean))         + str(bmi mean)
In [64]:	<pre>print('BMI standard deviation: ' + str(bmi_std)) # mean is now bigger, std is now less  BMI mean: 32.46995884773663 BMI standard deviation: 6.885098188024897  # 13.9. By using the Chebysheffs Theorem infer what is the range for the BMI variable for 75% of the people : # What is the range for 89% of the people in your dataset. # 1) mean - (lower boundary) = within number # 2) mean - (upper boundary) = within number # k - number of standard deviations # k = the within number / std # wn = sqrt(1/(1-x))* std; x = percentage wn75 = np.sqrt(1/(1 - 75/100)) * bmi_std wn89 = np.sqrt(1/(1 - 89/100)) * bmi_std print(wn75)</pre>
In [65]:	print('range for 75% is between '+ str(bmi_mean - wn75) + ' and ' + str(bmi_mean + wn75)) print(wn89) print('range for 89% is between '+ str(bmi_mean - wn89) + ' and ' + str(bmi_mean + wn89)) # as a result within numbers are smaller, the range also became smaller because zeros seemed to be just outl.  13.770196376049794 range for 75% is between 18.69976247168683 and 46.240155223786424 20.75935212221311 range for 89% is between 11.710606725523519 and 53.229310969949736  # 13.10. Create a histogram for the BMI variable, plot the mean and meadian on the histogram to answer the standard in the BMI distribution skewed based on the position of the mean versus the median? # b) Calculate the skewness of BMI # c) Based on the value of BMI median and on the chart below infer what percentage of people in your dataset.
	# create histogram plt.hist(dataset.BMI) plt.title('BMI histogram') plt.xlabel('BMI') plt.ylabel('Amount of people') bmi_median = np.median(dataset.BMI) print('BMI median = '+str(bmi_median)) plt.axvline(bmi_mean, color='yellow', linestyle='-') plt.axvline(bmi_median, color='red', linestyle='') plt.rcParams['figure.figsize'] = [10, 5] plt.show() bmi_skewness = dataset.BMI.skew() print('BMI skewness = '+str(bmi_skewness)) # c: # obese people are with bmi = 30 or higher print('there is this exact percantage of people with obesity: '+ str(len(dataset.BMI[dataset.BMI >= 30])/(len # skewness somehow is now positive and bigger, obesity without those zeroes also appears more often # mean and median differ more Now # how can skewness be bigger when i removed the outliers? # also they are no longer on the highest bar, which is also weird for median I guess  BMI median = 32.4  BMI histogram
	175 - 150 - 100 - 100 - 25 - 25 - 26 - 27 - 28 - 29 - 20 30 40 50 60
In [66]:	BMI skewness = 0.5951875009633939 there is this exact percantage of people with obesity: 62.55144032921811  # 13.11. doing step 10 for blood pressure plt.hist(dataset.BloodPressure) plt.title('Blood pressure histogram') plt.xlabel('Blood pressure') plt.ylabel('Amount of people') blood pressure_median = np.median(dataset.BloodPressure) blood pressure_median = np.mean(dataset.BloodPressure) print('Blood pressure median = '+str(blood pressure_median)) plt.axvline(blood_pressure_mean, color='yellow', linestyle='-') plt.axvline(blood_pressure_mean, color='red', linestyle='') plt.rcParams['figure.figsize'] = [10, 5] plt.show() # a: distribution is skewed on the mean more # b: calculate skewness, it is positive and also very small, therefore tails are on the right, so we should if skewness is pretty small here blood_pressure_skewness = dataset.BloodPressure.skew() print('Blood_pressure_skewness = '+str(blood_pressure_skewness)) # c: obesity is not the thing here
	Blood pressure median = 72.0  Blood pressure histogram  200  9150  50
	0 20 40 60 80 100 120 Blood pressure